

## Convolutional neural network approach to event position reconstruction in DarkSide-50 experiment

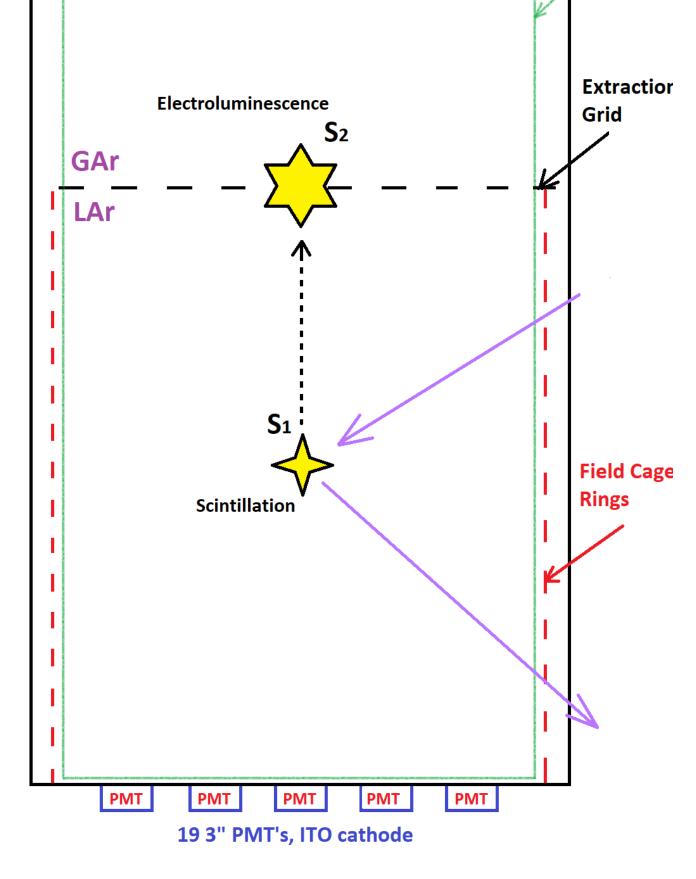
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## Abstract

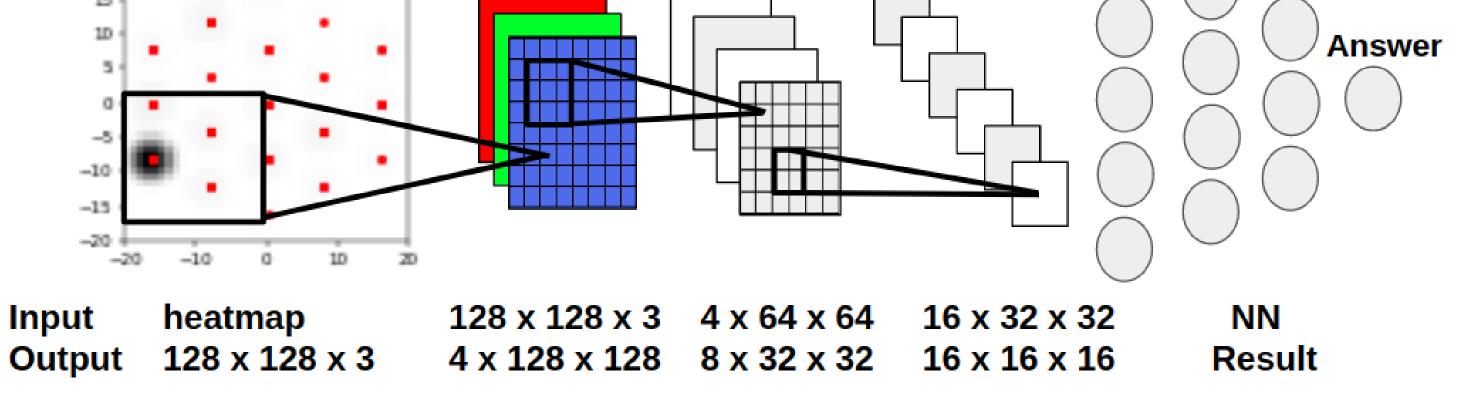
Neural networks are currently used in various fields of science, technology, as well as in experiments related to particle physics. DarkSide-50 is a two-phase (liquid and gas) argon TPC which has two main signals: scintillation in LAr and electroluminescence in GAr. Currently, only the more energetic second signal is used for position reconstruction. However, the used reconstruction method significantly reduces the fiducial volume of the detector. We use convolutional neural network (CNN) technique from Keras Python package to create new method of X-Y position reconstruction in DarkSide-50 experiment that will be equally effective and possibly outperform the existing one. We create a heatmap of each of the  $15 \cdot 10^3$  Monte Carlo simulation events, divide them into the train and test samples, train our model and improve result by tuning model parameters.

| Introduction to DarkSide-50 experiment  | CNN technique                             |
|---|---|
| 19 3" PMT's, ITO anode TPB • $(46.4 \pm 0.7)$ kg. of LAr in the Time PMT PMT PMT PMT TIME TIME Projection Chamber(TPC); | Conv1 —> Pool1 —> Conv2 —> Pool2 —> Conv3 |



• 90% CL upper limit on the WIMP-nucleon spinindependent cross section of  $6.1 \times 10^{-44}$  cm<sup>2</sup> for a WIMP mass of 100 GeV/c<sup>2</sup>;

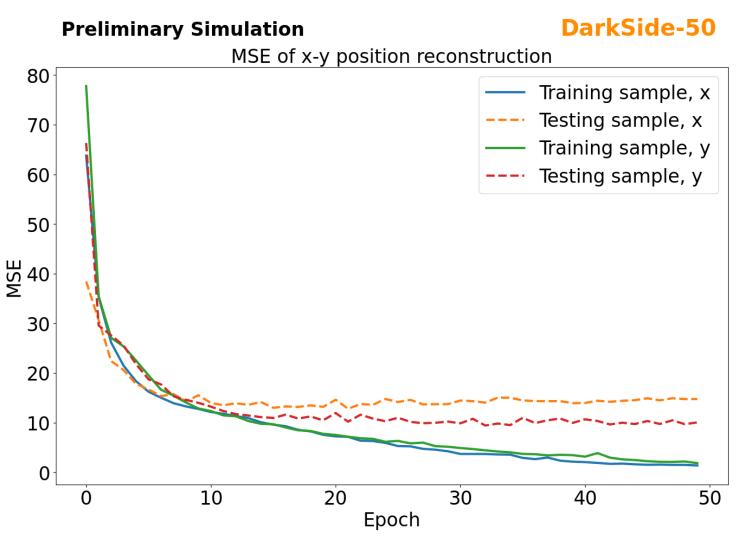
- Gas layer between the LAr surface and the TPC anode;
- Viewed by 38 PhotoMultiplier Tubes(PMTs), 19 each on the top and the bottom.;
- It has wavelength shifter tetraphenyl butadiene (TPB),  $128nm \rightarrow 420nm;$
- Vertical electric field in LAr to make electrons drift up;
- PSD: detector can exploit  $\frac{\text{scintillation}}{\text{ionization}}$  to reject  $\beta/\gamma$  back-ground;
- Water Cherenkov Detector (shielding and muon detection): 11 m-diameter, 10 m-high cylindrical tank, high purity water, 80 8" PMTs on the side and bottom of the water tank;
- Liquid Scintillator Veto (neutron and  $\gamma$ -ray rejections): 4.0 mdiameter stainless steel sphere, 30



- Heatmap 2D object that shows light pattern distribution in PMT's;
- Convolution layer per-

forms convolution of input layer with different kernels to detect leading features of the input; 60

- Max pooling layer downsamples the input representa- \$\$\frac{40}{240}\$ tion by taking the maximum 30
  value over the window defined 20
  by pool size for each dimension along the features axis; 0
- Fully connected layer NN which take the results of the convolution/pooling and use linear transform and activa-

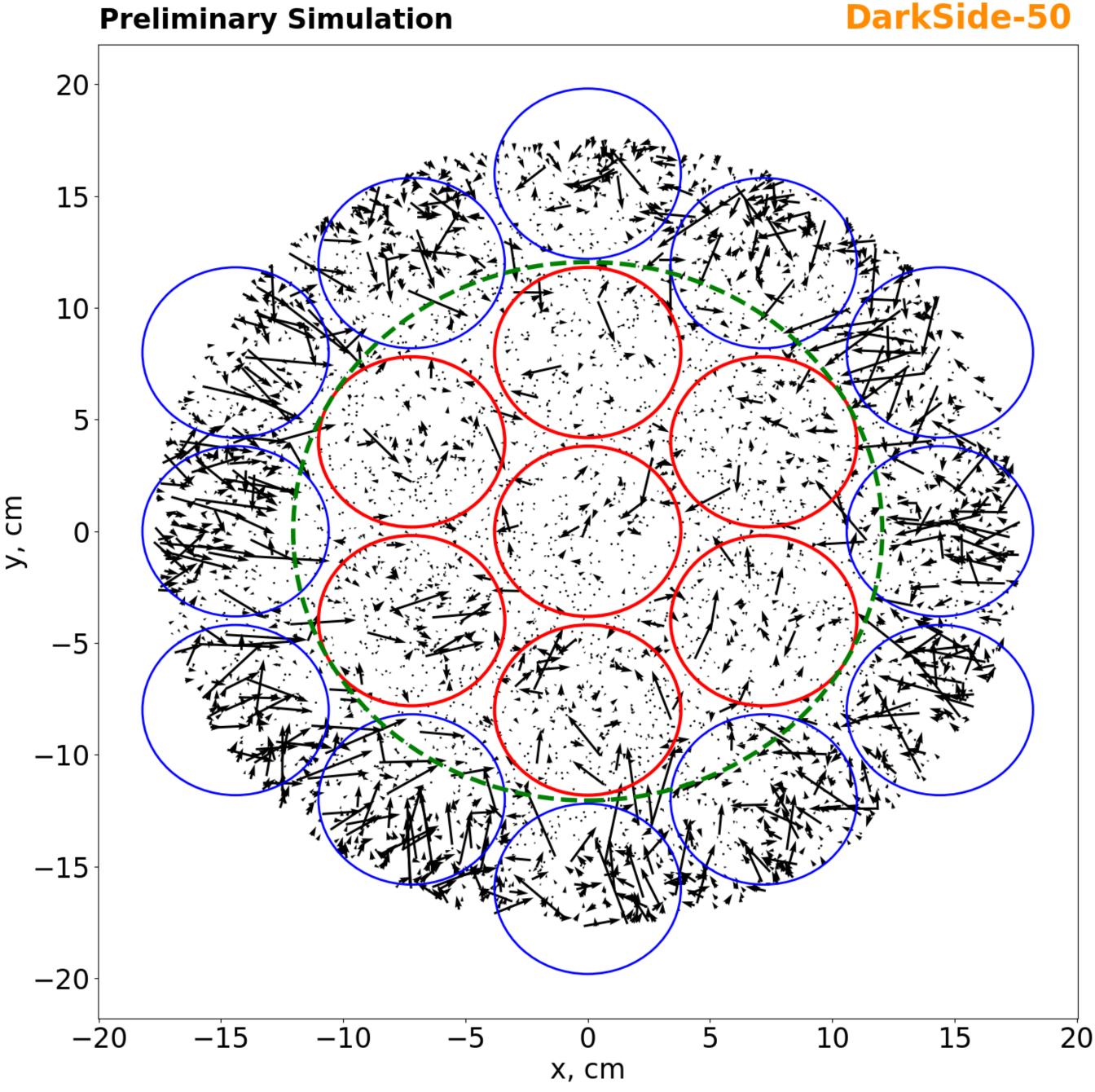


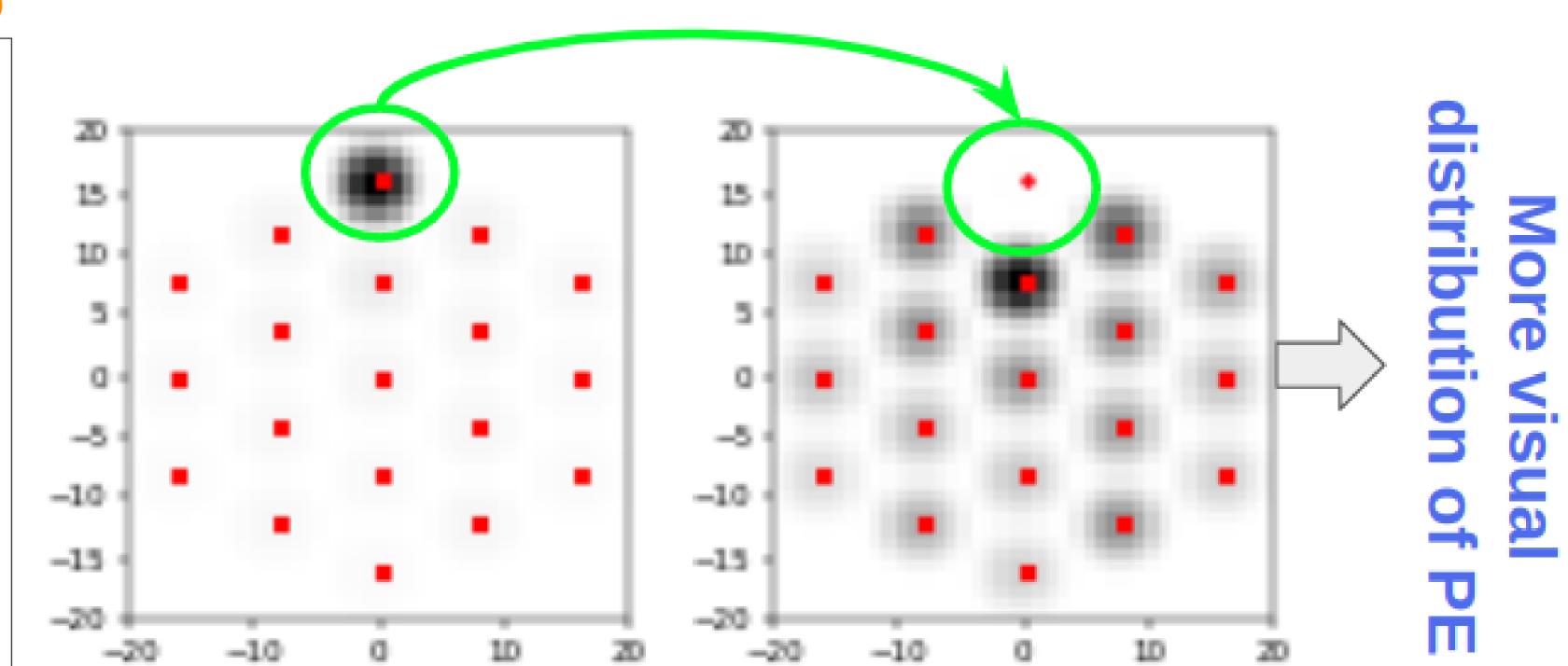
• Main goal – is to achieve the minimum value of the mean

• LNGS, Hall C at a depth of 3800 m.w.e.;

t of borated liquid scintillator, 110 8" PMTs. tion to predict X-Y coordinates. squared error (MSE) parameter

## Application CNN for event reconstruction





**Top figure**: Photoelectrons (PEs) distribution from  $S_2$  signal in top 19 PMT's: Left: original distribution;

Right: distribution in which the value of the PMT, which detected most PEs, is set to

one.

**Left figure**: Plot, that shows reconstruction error: arrowtail and arrowhead correspond to the original and reconstructed position of the event:

- **red circles** correspond to 7 internal PMT's;
- **green circle** corresponds to the volume of the detector in which the error was calculated;
- Total MSE on train sample ≈ 4cm, total validation MSE ≈ 11cm for both X and Y reconstruction.

## Summary

- Convolutional Neural Network works and can reconstruct X-Y position of events in DarkSide-50 experiment;
- MSE on train sample for 7 central PMT's  $\approx$  3 cm and 2.5 cm, validation MSE  $\approx$  4 cm and 3 cm for reconstruction of X and Y coordinates respectively;
- Next steps are: **increase** event statistic, **tune** CNN structure, **construct** new types of heatmaps.